- ¹ Title: Tree height and hydraulic traits shape growth responses across droughts in a temperate broadleaf
- 2 forest

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Summary

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- As climate change is driving increased drought frequency and severity in many forested regions around 23 the world, mechanistic understanding of the factors conferring drought resistance in trees is 24 increasingly important. The dendrochronological record provides a window through which we can 25 understand how tree size and species' traits shape tree growth responses during droughts. 26
- We analyzed tree-ring records for twelve species that comprise 97% of the woody productivity of the 27 25.6-ha ForestGEO plot in a broadleaf deciduous forest of northern Virginia (USA) to test hypotheses 28 on how tree height, microenvironment characteristics, and species' traits shaped drought responses 29 across the three strongest regional droughts over a 60-year period (1950 - 2009). 30
- Individual-level drought resistance decreased with tree height and was not significantly linked to canopy position. The potentially greater rooting volume of larger trees did not confer an advantage in 32 sites with low topgraphic wetness index. Resistance was greater among species whose leaves 33 experienced less shrinkage upon desiccation and lost turgor (wilted) at more negative water potentials.
- We conclude that tree height and hydraulic traits influence growth responses during drought, as 35 recorded in the tree-ring record spanning historical droughts. Thus, these factors can be useful for 36 predicting future drought responses under climate change. 37
- Key words: annual growth; canopy position; drought; Forest Global Earth Observatory (ForestGEO); leaf
- hydraulic traits; temperate broadleaf deciduous forest; tree height; tree-ring

40 Introduction

uncertainty as to how the terrestrial carbon sink, which is dominated by forests, will respond to climate change (Friedlingstein et al., 2006). An important aspect of this uncertainty lies with physiological responses 43 of trees to drought (Kennedy et al., 2019). In many forested regions around the world, the risk of severe 44 drought is increasing (Trenberth et al., 2014; Dai et al., 2018), often despite increasing precipitation (Intergovernmental Panel on Climate Change, 2015; Cook et al., 2015). Droughts, intensified by climate change, have been affecting forests worldwide and are expected to continue as one of the most important drivers of forest change in the future (Allen et al., 2010, 2015). Understanding forest responses to drought requires elucidation of how tree size, microenvironment, and species' traits jointly influence individual-level drought resistance, and the extent to which their influence is consistent across droughts. However, it has proven difficult to resolve the many factors affecting tree growth during drought with available forest census data, which only rarely captures extreme drought, and with tree-ring records, which capture multiple droughts but rarely consider the roles of tree size and microenvironment. 53 Many studies have shown that within species, large trees tend to be more affected by drought. Greater growth reductions for larger trees was first shown on a global scale by Bennett et al. (2015), and subsequent studies have reinforced this finding (e.g., Stovall et al. (2019); Hacket-Pain et al. (2016)). It has yet to be resolved which of several potential underlying mechanisms most strongly shape size trends in drought 57 response. First, tree height may be a primary driver. Taller trees face the biophysical challenge of lifting water greater distances against the effects of gravity and friction (McDowell et al., 2011; McDowell and Allen, 2015; Ryan et al., 2006; Couvreur et al., 2018). Vertical gradients in stem and leaf traits-including smaller and thicker leaves (higher leaf mass per area, LMA), greater resistance to hydraulic dysfunction (i.e., more 61 negative water potential at 50% loss of hydraulic conductivity, more negative P50), and lower hydraulic 62 conductivity at greater heights (Couvreur et al., 2018; Koike et al., 2001; McDowell et al., 2011)-enable trees 63 to become tall (Couvreur et al., 2018). Indeed, tall trees require xylem of greater hydraulic efficiency, such that xylem conduit diameters are wider in the basal portions of taller trees, both within and across species (Olson et al., 2018; Liu et al., 2019), and throughout the conductive systems of angiosperms (Zak et al. 2010, Olson et al. 2014,2018). Wider xylem conduits plausibly make large trees more vulnerable to embolism during drought (Olson et al., 2018), and traits conducive to efficient water transport may also lead to poor ability to recover from or re-route water around embolisms (Roskilly et al., 2019). Second, larger trees may have lower drought resistance because they tend to occupy more exposed canopy positions, where they experience higher solar radiation, greater wind speeds, and lower relative humidity (REFS-KAT). 71 Subcanopy trees tend to fare better specifically due to the benefits of a buffered environment (Pretzsch et al., 72 2018). Third, large trees tend to have larger root systems, which potentially counteracts some of the biophysical challenges they face by allowing greater access to water; however, it appears that this effect is 74 usually insufficient to offset the costs of height and/or crown exposure. Finally, tree size-related responses to drought can be modified by species' traits and their distribution across size classes (Meakem et al., 2018; Liu et al., 2019). Understanding the mechanisms driving the greater relative growth reductions of larger trees 77 during drought will require sorting out the interactive effects of height, canopy position, root water access, and species' traits. Debates have also arisen regarding the traits influencing tree growth responses to drought. It has been

Forests play a critical global role in climate regulation (Bonan, 2008), yet there remains enormous

observed that ring-porous species showing higher drought tolerance than diffuse-porous species (Friedrichs

et al., 2009; Elliott et al., 2015; Kannenberg et al., 2019), but this classification does not resolve differences among the many species within each category. Commonly-measured traits including wood density and leaf mass per area (LMA) have been linked to drought responses in some temperate deciduous forests (Abrams, 1990; Guerfel et al., 2009; Hoffmann et al., 2011; Martin-Benito and Pederson, 2015) and other forest biomes around the world (Greenwood et al., 2017). However, in other cases these traits could not explain drought tolerance (Maréchaux et al., 2019), or the direction of response was not always consistent. For instance, 87 higher wood density has been associated with greater drought resistance at a global scale (Greenwood et al., 2017), but it correlated negatively with tree performance during drought in a broadleaf deciduous forest in the southeastern United States (Hoffmann et al., 2011). Thus, the perceived influence of these traits on drought resistance may actually reflect indirect correlations with other traits that more directly drive drought responses (Hoffmann et al., 2011). Recent work has shown a great potential for hydraulic traits to predict growth and mortality responses. Hydraulic traits including water potentials at which percent loss of conductivity surpass a certain threshold (P50, P80, P88) and hydraulic safety margin correlate with drought performance (Anderegg et al., 2018) but are time-consuming to measure and therefore infeasible for predicting or modeling drought responses in highly diverse forests (e.q., in the tropics). More easily measurable leaf hydraulic traits with direct linkage to plant hydraulic function can explain greater variation in plant distribution and function (Medeiros et al., 2019). These include leaf area shrinkage upon desiccation (PLA_{dry}) (Scoffoni et al., 2014) and the leaf water potential at turgor loss point (π_{tlp}) , i.e., the water qq potential at which leaf wilting occurs (Bartlett et al., 2016). The abilities of both PLA_{dry} and π_{tlp} to 100 explain tree performance under drought remains untested. Here, we examine how tree height, microenvironment characteristics, and species' traits collectively shape 102 drought responses. We test a series of hypotheses and associated specific predictions (Table 1) based on the 103

combination of tree-ring records from three droughts (1966, 1977, 1999), species functional and hydraulic 104 trait measurements, and census data from a large forest dynamics plot in Virginia, USA. First, we focus on the role of tree height and its interaction with microenvironment. We test hypotheses designed to disentangle 106 the relative importance of tree height; crown exposure; and soil water availability, which should be greater for 107 larger trees in dry but not in perpetually wet microsites. Second, we focus on the role of species' functional and hydraulic traits, testing the hypothesis that species' traits-particularly leaf hydraulic traits-predict Rt. 109 We test predictions that drought resistance is higher in ring-porous than semi-ring and diffuse-porous species, 110 that it is correlated with wood density-either postively (Greenwood et al., 2017) or negatively (Hoffmann 111 et al., 2011)– and positively correlated with LMA, and that hydraulic leaf traits including PLA_{dry} and π_{tlp} 112 are better predictors. 113

114 Materials and Methods

115 Study site

Research was conducted at the 25.6-ha ForestGEO (Forest Global Earth Observatory) study plot at the
Smithsonian Conservation Biology Institute (SCBI) in Virginia, USA (38°53'36.6"N, 78°08'43.4"W) (Bourg
et al., 2013; Anderson-Teixeira et al., 2015a). SCBI is located in the central Appalachian Mountains near the
northern boundary of Shenandoah National Park. Elevations range from 273 to 338 m above sea level with a
topographic relief of 65m (Bourg et al., 2013). Climate is humid temperate, with mean annual temperature
of 12.7°C and precipitation of 1005 mm during our study period (1960-2009; source: CRU TS v.4.01; Harris
et al. (2014)). Dominant tree taxa within this secondary forest include Liriodendron tulipifera, oaks (Quercus

spp.), and hickories (Carya spp.).

Data collection and preparation 124

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Within or just outside the ForestGEO plot, we collected data on a suite of variables including tree size, 125 microenvironment characteristics, and species traits (Table 2). The SCBI ForestGEO plot was censused in 2008, 2013, and 2018 following standard ForestGEO protocols, whereby all free-standing woody stems > 1cm 127 diameter at breast height (DBH) were mapped, tagged, measured at DBH, and identified to species (Condit, 128 1998). From this census data, we used measurements of DBH from 2008 to calculate historical DBH and data for all stems \geq 10cm to analyze functional trait composition relative to tree height (all analyses described 130 below). Census data are available through the ForestGEO data portal (www.forestgeo.si.edu). 131

We analyzed tree-ring data (xylem growth increment) from 571 trees representing the twelve species with the 132 greatest contributions to woody aboveground net primary productivity $(ANPP_{stem})$, which together 133 comprised 97% of study plot $ANPP_{stem}$ between 2008 and 2013 (Helcoski et al., 2019) (Fig. S1). Cores (one 134 per tree) were collected within the ForestGEO plot at breast height (1.3m) in 2010-2011 or 2016-2017. In 135 2010-2011, cores were collected from randomly selected live trees of each species that had at least 30 136 individuals > 10 cm DBH (Bourg et al., 2013). In 2016-2017, cores were collected from all trees found dead 137 during annual mortality censuses (Gonzalez-Akre et al., 2016). Cores were sanded, measured, and crossdated using standard procedures, as detailed in (Helcoski et al., 2019). The resulting chronologies were published in Zenodo (DOI: 10.5281/zenodo.2649302) in association with Helcoski et al. (2019).

For each cored tree, we combined tree-ring records and allometric equations of bark thickness to reconstruct 141 DBH for the years 1950-2009. Prior DBH was estimated using the following equation:

$$DBH_Y = DBH_{2008} - 2 * \left[\sum_{y=ar=Y}^{2008} (r_{ring,Y}) - r_{bark,Y} + r_{bark,2008} \right]$$

Here, Y denotes the year of interest, r_{ring} denotes ring width derived from cores, and r_{bark} denotes bark thickness. Bark thickness was estimated from species-specific allometries based on the bark thickness data

from the site (Anderson-Teixeira et al., 2015b). Specifically, we used linear regression on log-transformed data to relate r_{bark} to diameter inside bark from 2008 data (Table S1), which were then used to determine r_{bark} in the DBH reconstruction. 147 Tree heights (H) were measured by several researchers for a variety of purposes between 2012 to 2019 148 (n=1,518 trees). Measurement methods included direct measurements using a collapsible measurement rod 149 on small trees (NEON, 2018) or a tape measure on recently fallen trees (this study); geometric calculations using clinometer and tape measure (Stovall et al., 2018a) or digital rangefinders (Anderson-Teixeira et al., 151 2015b; NEON, 2018); and ground-based LiDAR (Stovall et al., 2018b). Rangefinders used either the tangent 152 method (Impulse 200LR, TruPulse 360R) or the sine method (Nikon ForestryPro) for calculating heights. Both methods are associated with some error (Larjavaara and Muller-Landau, 2013), but in this instance 154 there was no clear advantage of one or the other. Measurements from the National Ecological Observatory 155 Network (NEON) were collected near the ForestGEO plot following standard NEON protocol, whereby vegetation of short stature was measured with a collapsible measurement rod, and taller trees with a 157 rangefinder (NEON, 2018). Species-specific height allometries were developed (Table S2) using logarithmic regression $(ln[H] \sim ln[DBH])$. For species with insufficient height data to create reliable species-specific

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allometries (n=2, JUNI and FRAM), heights were calculated from an equation developed by combining the
    height measurements across all species. We then used these allometries to estimate H for each drought year,
    Y, based on reconstructed DBH_Y.
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    Crown position—a categorical variable including dominant, co-dominant, intermediate, and suppressed—was
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    recorded for all cored trees that remained standing during the growing season of 2018 following the protocol
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    of Jennings et al. (1999). While some tree crowns undoubtedly changed position over the past several
    decades, in this case the bias would be unlikely to result in false acceptance of the prediction that dominant
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    trees have the lowest Rt (i.e., type I error unlikely, type II error possible), making our hypothesis test
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    conservative. An analysis of crown position relative to height (Fig. 2d) and height changes since the
    beginning of the study period indicated that changes between focal drought years (1966, 1977, and 1999; see
169
    below) were fairly small relative to differences among canopy positions (Fig. S3), with average tree height
170
    growth confined to ~0.82 m from 1966 to 1977, ~1.45 m from 1977 to 1999, and ~1.97 m from 1999 to 2018.
    However, dominant and co-dominant trees were similar in height (Figs. 2d, S3).
172
    Topographic wetness index (TWI) was calculated using the dynatopmodel package in R (Fig. S1) (?).
173
    Originally developed by Beven and Kirkby (1979), TWI was part of a hydrological run-off model and has
174
    since been used for a number of purposes in hydrology and ecology (Sørensen et al., 2006). TWI calculation
175
    depends on an input of a digital elevation model (DEM; ~3.7 m resolution from the elevatr package (?)), and
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    from this yields a quantitative assessment defined by how "wet" an area is, based on areas where run-off is
177
    more likely. From our observations in the plot, TWI performed better at categorizing wet areas than the
178
    Euclidean distance from the stream.
179
    Hydraulic traits were collected in August 2018 (Table 3). We sampled small sun-exposed branches up to
180
    eight meters above ground from three individuals of each species in and around the ForestGEO plot.
181
    Sampled branches were re-cut under water at least two nodes above the original cut and re-hydrated
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    overnight in covered buckets under opaque plastic bags before measurements were taken. Rehydrated leaves
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    taken towards the apical end of the branch (n=3 per individual: small, medium, and large) were scanned,
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    weighed, dried at 60^{\circ} C for \geq 48 hours, and then re-scanned and weighed. Leaf area was calculated from
185
    scanned images using the LeafArea R package (Katabuchi, 2019). LMA was calculated as the ratio of leaf
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    dry mass to fresh area. PLA_{dry} was calculated as the percent loss of area between fresh and dry leaves.
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    Wood density was calculated for ~1cm diameter stem samples (bark and pith removed) as the ratio of dry
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    weight to fresh volume, which was estimated using Archimedes' displacement. We used the rapid
189
    determination method of Bartlett et al. (2012) to estimate osmotic potential at turgor loss point (\pi_{tlp}).
    Briefly, two 4 mm diameter leaf discs were cut from each leaf, tightly wrapped in foil, submerged in liquid
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    nitrogen, perforated 10-15 times with a dissection needle, and then measured using a vapour pressure
192
    osmometer (VAPRO 5520, Wescor, Logan, UT, USA). Osmotic potential (\pi_{osm}) given by the osmometer was
    used to estimate (\pi_{tlp}) using the equation \pi_{tlp} = 0.832\pi_{osm}^{-0.631} (Bartlett et al., 2012).
194
    To characterize how environmental conditions vary with height, data were obtained from the NEON tower
195
    located <1km from the study area via the neon Utilities package (?). We used wind speed, relative humidity,
196
    and air temperature data, all measured over a vertical profile spanning heights from 7.2 m to above the top
    of the tree canopy (31.0 or 51.8m, depending on censor), for the years 2016-2018 (NEON, 2018). After
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    filtering for missing and outlier values, we determined the daily minima and maxima, which we then
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    aggregated at the monthly scale.
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Identifying drought years
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    We identified the three largest droughts within the time period 1950-2009, defining drought (Slette et al.,
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    2019) as events with both anomalously dry peak growing season climatic conditions. Specifically, we used the
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    metric of Palmer Drought Severity Index (PDSI) during May-August (MJJA; Table S3), which were
204
    identified by Helcoski et al. (2019) as the months of the current year to which annual tree growth was most
205
    sensitive at this site. PDSI divisional data for Northern Virginia were obtained from NOAA
    (https://www7.ncdc.noaa.gov/CDO/CDODivisionalSelect.jsp) in December 2017. Based on this, we
207
    identified three drought years - 1966, 1977, and 1999 (Figs. 1, S2, Table S3).
208
    The droughts differed in intensity and antecedent moisture conditions (Fig. S2, Table S3). The 1966 drought
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    was preceded by two years of moderate drought during the growing season and severe to extreme drought
210
    starting the previous fall and in August reached the lowest growing season PDSI (-4.82) of the three
211
    droughts. The 1977 drought was the least intense throughout the growing season, and it was preceded by 2.5
212
    years of near-normal conditions, making it the mildest of the three droughts. The 1999 drought was preceded
    by wetter than average conditions until the previous June, but reached the lowest PDSI during May-July
214
    (-4.53).
215
    Statistical Analysis
    For each drought year, we calculated drought resistance (Rt) as the ratio of basal area increment (BAI)
    during drought to the mean BAI over the five years preceding the drought (Lloret et al., 2011). Thus, Rt
218
    values <1 and >1 indicate growth reductions and increases, respectively. Because the Rt metric could be
219
    biased by directional pre-drought growth trends, we also tried an intervention time series analysis (ARIMA,
    (?)) that predicted mean drought-year growth based on trends over the past ten years and used this value in
221
    place of the five-year mean in calculations of resistance (Rt_{ARIMA}= observed BAI/ predicted BAI).
222
    Because Rt tended to produce more reasonable estimates than Rt_{ARIMA} when there was a large difference
    between these metrics, we selected Rt as our focal metric, presenting parallel results for Rt_{ARIMA} in the
224
    Supplementary Info. We focus exclusively on drought resistance (Rt or Rt_{ARIMA}), and not on the resilience
225
    metrics described in Lloret et al. (2011), because (1) we would expect resilience to be controlled by a
    different set of mechanisms, and (2) the findings of DeSoto et al. (2020) suggest that Rt is a more important
227
    drought response metric for angiosperms.
228
    Analyses focused on testing the predictions presented in Table 1, with Rt (or Rt_{ARIMA}) as the response
229
    variable. Models were run for all drought years combined and for each drought year individually. The general
230
    statistical model for hypothesis testing was a generalized linear mixed model (GLMM), implemented in the
231
    XX package in R [REF], with Rt (or Rt_{ARIMA}) as the response variable, tree nested within species as a
232
    random effect, and independent variables including drought year (multi-drought model only), ln[H]^*TWI,
233
    crown position, and 1-2 species traits (see below). We used AICc to assess model selection, and
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    conditional/marginal R-squared to assess model fit as implemented in the AICcmodavg package in R (?).
235
    AICc refers to a corrected version of AICc, and is best suited for small data sizes (see Brewer et al., 2016).
    To avoid over-fitting models with five species traits (Table 2) across only 12 species, we did not include all
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    traits as fixed effects in a single GLMM, but rather conducted individual tests of each species trait to
238
    determine the relative importance and appropriateness for inclusion in the main model. These tests followed
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    the model structure specified above. Trait variables were considered appropriate for inclusion in the main
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    model if they had a consistent direction of response across all droughts and if their addition to a
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corresponding null model lacking the trait improved fit (at $\Delta AICc \geq 1.0$) in at least one drought year (Table 4). We note that the $\Delta AICc \geq 1.0$ criterion is not a test of significance, but of whether the variable has enough influence to be considered as a *candidate* variable in full models.

We then determined the top full models for predicting Rt (or Rt_{ARIMA}). To do so, we compared models with all possible combinations of candidate variables and identified the full set of models within $\Delta AICc=2$ of the best model (that with lowest AICc). When a variable appeared in all of these models and the sign of the coefficient was consistent across models, we viewed this as support for the acceptance/rejection of the associated prediction (Table 1). If the variable appeared in some but not all of these models, and its sign we was consistent across models, we considered this partial support/rejection. In presentation of the results below, we note instances where the Rt_{ARIMA} model disagreed with the Rt model, but otherwise do not discuss the Rt_{ARIMA} model.

All analysis beyond basic data collection was performed using R version 3.5.3 (R Core Team, 2020). Other
R-packages aside from those already listed were very helpful in conducting analyses. These are listed in the
Supplementary Information. All data, code, and results are available through the SCBI-ForestGEO
organization on GitHub (https://github.com/SCBI-ForestGEO: SCBI-ForestGEO-Data and
McGregor_climate-sensitivity-variation repositories), with static versions corresponding to data and analyses
presented here archived in Zenodo (DOIs: 10.5281/zenodo.3604993 and [TBD], respectively.

259 Results

260 Community-level and species' drought responses

At the community level, cored trees showed substantial growth reductions in all three droughts, with a mean Rt of 0.86 in 1966 and 1999, and 0.84 in 1977 (Fig. 1b). Across the entire study period (1950-2009), the focal drought years were the three years with the largest fraction of trees exhibiting $Rt \le 0.7$. Specifically, in each drought, roughly 30% of the cored trees had growth reductions of $\ge 30\%$ ($Rt \le 0.7$): 29% in 1966, 32% in 1977, and 27% in 1999. However, some individuals exhibited increased growth, *i.e.*, Rt > 1.0: 26% of trees in 1966, 22% in 1977, and 26% in 1999.

Responses varied across species and by drought (Fig. 2). Averaged across all droughts, Rt was lowest in Liriodendron tulipifera (mean Rt = 0.66) and highest in Fagus grandifolia (mean Rt = 0.99).

²⁶⁹ Tree size, microenvironment, and drought resistance

Taller trees showed stronger growth reductions during drought when evaluating the three drought years together and for 1966 individually (Table 1; Fig. 4;). Specifically, ln[H] appeared, with negative coefficient, in the best models and all models within $\Delta AICc=2$ of these (Tables S6-S7). For the 1977 and 1999 droughts, ln[H] did not appear in the best models, but was included, with negative coefficient, among some of the top models, which were statistically indistinguishable ($\Delta AICc<2$) from the top model (Tables 1, S6-S7).

Crown position varied as expected with height (dominant > co-dominant > intermediate > suppressed), but with substantial variation (Fig. 3d). Crown position was never statistically significant (Table 1). It did, however, come out in the best models for 1977 and 1999 (Fig. 4), and was included in some of the top models for 1966 and all droughts combined (Table S6). When Rt_{ARIMA} was used as the predictor variable, crown position was never in a best model, and was included among the top models only for 1977. Canopy position did not have a consistent influence, when included in top models, across years: dominant trees had the lowest

In the years for which we have vertical profiles in climate data (2016-2018), taller trees-or those in dominant 282 crown positions- were generally exposed to higher evaporative demand during the peak growing season 283 months (May-August; Fig. 3). Specifically, maximum daily wind speeds were significantly higher above the 284 top of the canopy (40-50m) than within and below (10-30m) (Fig. 3a). Relative humidity was also somewhat 285 lower during June-August, ranging from ~50-80% above the canopy and ~60-90% in the understory (Fig. 3b). Air temperature did not vary across the vertical profile (Fig. 3c). 287 Rt had a significantly negative response to ln[TWI] in all drought years combined and in 1977, and a 288 negative effect of ln[TWI] was included in some of the models in 1999 and the Rt_{ARIMA} model for 1966 289 (Fig. 4, Tables 4-5). This negates the idea that trees in moist microsites would be less affected by drought. 290 Nevertheless, we tested for a negative ln[H] * ln[TWI] interaction, which could indicate that smaller trees 291 (with smaller rooting volume) are more susceptible to drought in drier microenvironments with a deeper 292 water table. This hypothesis was rejected, as the ln[H] * ln[TWI] interaction was never the significant, and had a positive sign in any top models in which it appeared (Tables 1, S6-S7). This term did appear, with 294 positive coefficient, in the best Rt_{ARIMA} model for all years combined (Table S7). 295 Species' traits and drought resistance Wood density, LMA, and xylem porosity were all poor predictors of Rt (Tables 1,S4-S5). Wood density and LMA were never significantly associated with Rt in the single-variable tests and were therefore excluded 298 from the full models. Xylem porosity was also excluded from the full models, as it had no significant 299 influence for all droughts combined and had contrasting effects in the individual droughts: whereas ring-porous species had higher Rt than diffuse- and semi-ring- porous species in the 1966 and 1999 droughts, 301 they had lower Rt in 1977 (Table S4). It is noteworthy that the two diffuse-porous species in our study, 302 Liriodendron tulipifera and Fagus grandifolia, were at opposite ends of the Rt spectrum (Fig. 2), further refuting the idea that xylem porosity is a useful predictor of Rt at this site. 304 In contrast, PLA_{dry} , and π_{tlp} were linked to drought responses (Fig. 4; Tables 1,S4-S7). Both had consistent 305 signs across all droughts and explained modest amounts of variation ($\Delta AICc > 1.0$) during at least one of 306 the three droughts (Table S4), qualifying them as candidate variables for the full model. PLA_{dry} had a 307 significant influence, with negative coefficient, in full models for the three droughts combined and for the 308 1966 drought individually (Fig. 4; Tables S6-S7). For 1977 and 1999, it was included, with negative 309 coefficient, in some of the top models (Tables S6-S7). π_{tlp} was included, with negative coefficient, in the best 310 model for all droughts combined and for the 1977 drought individually (Fig. 4; Table 5). It was included in 311

Rt (or Rt_{ARIMA}) in 1977, but in all other years suppressed trees had the lowest Rt (Fig. 4; Tables S6-S7).

313 Discussion

312

some of the top models for 1999 (Tables S6-S7).

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Tree height, microenvironment, and hydraulic traits shaped tree growth responses across three droughts at our study site (Table 1, Fig. 4). The greater susceptibility of larger trees to drought, similar to forests worldwide (Bennett et al., 2015), was driven primarily by their height (Stovall et al., 2019). We found only a marginal additional effect of crown exposure, with a tendency for lowest Rt among the most exposed (dominant) and suppressed trees. The negative effect of height on Rt held after accounting for species' traits. There was no evidence that soil water availability increased drought resistance; in contrast, trees in wetter topographic positions had lower Rt (Zuleta et al., 2017; Stovall et al., 2019), and the larger potential rooting

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volume of large trees provided no advantage in the drier microenvironments. Drought resistance was not
321
    linked to species' LMA, wood density, or xylem type (ring- vs. diffuse porous), but was negatively correlated
    with leaf hydraulic traits (PLA_{dry}, \pi_{tlp}). This is the first report to our knowledge linking PLA_{dry} and \pi_{tlp}
323
    to growth reduction during drought. The directions of these responses were consistent across droughts,
324
    supporting the premise that they were driven by fundamental physiological mechanisms. However, the
    strengths of each predictor varied across droughts (Fig. 4; Tables S6-S7), indicating that drought
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    characteristics interact with tree size, microenvironment, and traits to shape which individuals are most
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    affected. These findings advance our knowledge of the factors that make trees vulnerable to growth declines
    during drought-and, by extension, likely make them more vulnerable to mortality (Sapes et al., 2019).
329
    The droughts considered here were of a magnitude that has occurred with an average frequency of
330
    approximately once every 10-15 years (Fig. 1a, Helcoski et al. (2019)) and had substantial but not
331
    devastating impacts on tree growth (Figs. 1b, 2). These droughts were classified as severe (1977) or extreme
    (1966, 1999) according to the PDSI metric and have been linked to tree mortality in the eastern United
333
    States (Druckenbrod et al., 2019); however, extreme, multiannual droughts or so-called "megadroughts" of
334
    the type that have triggered massive tree die-off in other regions (e.g., Allen et al. (2010); Stovall et al.
335
    (2019)) have not occurred in the Eastern United States within the past several decades (Clark et al., 2016).
336
    Of the droughts considered here, the 1966 drought, which was preceded by two years of dry conditions (Fig.
337
    S2), severely stressed a larger portion of trees (Fig. 1b). The tendency for large trees to have lowest
338
    resistance was most pronounced in this drought, consistent with other findings that this physiological
339
    response increases with drought severity (Bennett et al., 2015; Stovall et al., 2019). Across all three droughts,
340
    the majority of trees experienced reduced growth, but a substantial portion had increased growth (Fig. 1b),
341
    potentially due to decreased leaf area of competitors during the drought (REF-if we can find one), and
342
    consistent with prior observations that smaller trees can exhibit increased growth rates during drought
343
    (Bennett et al., 2015). It is likely because of the moderate impact of these droughts, along with other factors
    influencing tree growth (e.g., stand dynamics), that our best models characterize only a modest amount of
345
    variation in Rt: 11-12% for all droughts combined, and 18-26% for each individual drought (Table S6).
346
    Our analysis indicates that tree height has a stronger influence on drought response than does canopy
347
    position (Fig. 4; Tables 1, S6-S7). This is consistent with, and reinforces, previous findings that biophysical
348
    constraints make it impossible for trees to efficiently transport water to great heights and simultaneously
349
    maintain strong resistance and resilience to drought-induced embolism (Olson et al., 2018; Couvreur et al.,
350
    2018; Roskilly et al., 2019). However, the collinearity between the two variables (Fig. 3d) makes it
351
    impossible to confidently partition causality. Taller trees are more likely to be in dominant canopy positions
352
    (Fig. 3d) and, largely as a consequence of their position relative to others, face different microenvironments
353
    (Fig. 3a-b). Even under non-drought conditions, evaporative demand and maximum leaf temperatures
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    increase with tree height (Smith and Nobel, 1977; Bretfeld et al., 2018; Kunert et al., 2017), and such
355
    conditions would incur additional stress during drought, when solar radiation tends to be higher and less
    water is available for evaporative cooling of the leaves (Campbell & Norman REF). However, some
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    decoupling between height and canopy position is introduced by the configuration of neighboring trees (Fig.
358
    3d) (Muller-Landau et al., 2006), and height was an overall stronger predictor of drought response than
359
    crown position (Fig. 4; Tables 1, S6-S7). Belowground, taller trees would tend to have larger root systems,
360
    but the potentially greater access to water did not override the disadvantage conferred by height-and, in fact,
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    greater moisture access in non-drought years (here, higher TWI) appears to make trees more sensitive to
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drought (Zuleta et al., 2017; Stovall et al., 2019). 363 Our analysis has the limitation that canopy positions were recorded in 2018, as opposed to the years of the 364 droughts. However, because trees would generally advance towards more dominant positions as they grow and as neighbors die, changing canopy positions would bias against the acceptance of our hypothesis. The 366 implication is that dominant crown positions did have a marginally negative influence on Rt, which makes 367 sense in light of the vertical environmental gradients described above and agrees with previous studies showing lower drought resistance in more exposed trees (Liu and Muller, 1993; Suarez et al., 2004; 369 Scharnweber et al., 2019). It is safe to assume that currently suppressed trees were suppressed throughout 370 our analysis period, and their relatively low Rt (after accounting for height effects) is real, perhaps as a result of competition (Sohn et al., 2016). The observed height-sensitivity of Rt, together with the lack of 372 advantage to large stature in drier topographic positions, agrees with the concept that physiological 373 limitations to transpiration under drought shift from soil water availability to the plant-atmosphere interface as forests age (Bretfeld et al., 2018), such that tall, dominant trees are the most sensitive in mature forests. 375 Additional research comparing drought responses of young and old forest stands, along with short and tall 376 isolated trees, would be valuable for more clearly disentangling the roles of tree height and crown exposure. 377 The development of tree-ring chronologies for the twelve most dominant tree species at our site (Helcoski 378 et al., 2019; Bourg et al., 2013) gave us the sample size to compare historical drought responses across 379 species (Fig. 2) and associated traits at a single site (see also Elliott et al., 2015). Concerted measurement of 380 leaf hydraulic traits of emerging importance (Scoffoni et al., 2014; Bartlett et al., 2016; Medeiros et al., 2019) 381 allowed novel insights into the role of hydraulic traits in shaping drought response. The finding that PLA_{dry} 382 and π_{tlp} can be useful for predicting drought responses of tree growth (Tables 1,4,5) is both novel and 383 consistent with previous studies linking these traits to habitat and drought tolerance. Previous studies have 384 demonstrated that π_{tlp} and PLA_{dry} are physiologically meaningful traits linked to species distribution along 385 moisture gradients (Maréchaux et al., 2015; Fletcher et al., 2018; Medeiros et al., 2019; Simeone et al., 2019; 386 Rosas et al., 2019), and our findings indicate that these traits also influence drought responses. Furthermore, 387 the observed linkage of π_{tlp} to Rt in this forest aligns with observations in the Amazon that π_{tlp} is higher in 388 drought-intolerant than drought-tolerant plant functional types and adds support to the idea that this trait 389 is useful for categorizing and representing species' drought responses in models (Powell et al., 2017). Because 390 both PLA_{dry} and π_{tlp} can be measured relatively easily (Bartlett et al., 2012; Scoffoni et al., 2014), they 391 hold promise for predicting drought growth responses across diverse forests. The importance of predicting drought responses from species traits increases with tree species diversity; whereas it is feasible to study 393 drought responses for all dominant species in most boreal and temperate forests (e.g., this study), this 394 becomes difficult to impossible for species that do not form annual rings, and for diverse tropical forests. 395 Although progress is being made for the tropics (Schöngart et al., 2017), a full linkage of hydraulic traits to 396 drought responses would be invaluable for forecasting how little-known species and whole forests will respond 397 to future droughts (Powell et al., 2017). As climate change drives increasing drought in many of the world's forests (Trenberth et al., 2014; Intergovernmental Panel on Climate Change, 2015), the fate of forests and their climate feedbacks will be 400 shaped by the biophysical and physiological drivers observed here. Large trees have been disproportionately 401 impacted by strong drought in forests around the world (Bennett et al., 2015; Stovall et al., 2019), and we show, at least at this site, that this is primarily driven by their height, potentially with some contributions 403 from canopy position. The distinction is important because it suggests that height per se makes trees

vulnerable, even if their crowns are somewhat protected by neighbors, whereas shorter solitary trees or the 405 dominant trees in young forests that recently established after logging or natural disturbances should be less vulnerable. This would suggest that, all else being equal, mature forests would be more vulnerable to 407 drought than young forests with short trees; however, root water access may limit the young forests (Bretfeld 408 et al., 2018), and species traits often shift as forests age. Early- to mid- successional species at our site (Liriodendron tulipifera, Quercus spp., Fraxinus americana) display a mix of traits conferring drought 410 tolerance and resistance (Table 3), and further research on how hydraulic traits and drought vulnerability 411 change over the course of succession would be valuable for addressing how drought tolerance changes as forests age (e.g. Rodríguez-Catón et al., 2015). In the meantime, the results of this study advance our 413 knowledge of the factors conferring drought resistance in a mature forest, opening the door for more accurate 414 forecasting of forest responses to future drought.

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426 Author Contribution

KAT, IM, and AJT designed the research. Tree-ring chronologies were developed by RH under guidance of AJT and NP. Trait data was collected by IM, JZ under guidance of NK and LS. Other plot data were collected by IM, AS, EGA, and NB under guidance of EGA and WM. Data analyses were performed by IM under guidance of KAT and VH. KAT and IM interpreted the results. IM and KAT wrote the first draft of manuscript, and all authors contributed to revisions.

432 Supplementary Information

- Table S1: Species-specific bark thickness regression equations
- Table S2: Species-specific height regression equations
- Table S3: Palmer drought severity index (PDSI) by month for focal droughts
- Figure S1: Map of ForestGEO plot showing TWI and location of cored trees
- Figure S2: Time series of Palmer Drought Severity Index (PDSI) for the 2.5 years prior to each focal drought
- Figure S3: Height (from reconstructed DBH) by canopy position across the three focal droughts and in the vear of measurement (2018)

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